

HC and LO Appendix - In-Silico Modeling of YCT-529

Capstone Project

Contents

LO Appendix	2
LO: #CS-Computationaltools	2
LO: #CS110-PythonProgramming	4
LO: #NS164-ResearchMethods	5
LO: #ns164-modelselection	7
HC Appendix	9
HC: #thesis	9
HC: #interventionalstudy	10
HC: #studyreplication	11
HC: #evidencebased	13
HC: #professionalism	14

LO Appendix

LO: #CS-Computationaltools

Description of the LO from the course syllabus: Use appropriate computational tools (e.g., SymPy, Sage, Python packages, modeling software) to analyze or solve problems. This LO emphasizes tool selection, proper implementation, and the effective communication of results. Mastery entails using computational solutions to manage complexity, validate data, and present findings clearly and accurately.

What is a “4” level application of this LO as described in the course? A **4-level** application involves correctly selecting and implementing computational tools while ensuring clear documentation, validation, and communication of results. This means integrating multiple tools to manage complexity, validating outputs, and effectively presenting findings. In pharmacokinetic modeling, this would involve configuring **PK-Sim** for simulations, using **Python** for data analysis, and ensuring consistency across outputs.

What would be a “4” level application of this LO in THIS project? A **4-level** application in this project would involve simulating **16,000 individual pharmacokinetic profiles** in **PK-Sim**, pre-processing large datasets in **Python**, and using advanced visualization techniques to compare inter-population variability, dietary effects, and drug-dose responses. **Cross-validation** of PK-Sim outputs with independent Python analyses would strengthen the computational framework.

Where is this LO applied in your project? This LO is applied throughout the project:

- **Methodology:** Details my use of PK-Sim for PBPK modeling and Python for statistical analysis.
- **Results:** Displays dose-response relationships, racial variability, and dietary condition effects.
- **Appendix C:** Documents how PK-Sim data was structured for Python analysis.
- **Appendix D:** Explains my coding workflow and key computational methods.

How have I applied this LO in THIS project? Evaluate the strength of my application: **Application 1: Multi-Platform Computational Workflow**

I integrated **PBPK modeling (PK-Sim)**, **large-scale data processing (Python)**, and **statistical visualization** to analyze YCT-529's pharmacokinetics. PK-Sim's built-in tools were insufficient for **cross-cohort dose-response analysis**, necessitating external computation and validation.

Application 2: Large-Scale Data Pre-Processing

With **16,000 individual simulations**, I wrote Python scripts to **aggregate, restructure, and convert units** for downstream analysis. PK-Sim could not directly compare **population-based trends**, requiring me to algorithmically compute **C_{max} distributions**, **dose-normalized plasma concentrations**, and inter-population variability.

Application 3: Validation and Advanced Visualization

I ensured **computational reproducibility** by **cross-validating PK-Sim results with Python-generated statistical summaries**. Additionally, I developed **custom visualization techniques** (dose-stratified C_{max} plots, overlay comparisons for placebo-drug response, and dietary condition sensitivity analyses) that **PK-Sim's built-in graphing tools** could not generate.

Consideration for a 5: Novel Application Beyond the Course Scope A **5-level** application extends beyond standard coursework expectations by demonstrating **innovation, computational complexity, and interdisciplinary expertise**. This project surpasses typical applications in three ways:

1. **Interdisciplinary Integration:** I established a workflow combining **PBPK modeling, statistical analysis, and custom visualization**, akin to professional pharmacometric research.
2. **Independent Model Refinement:** Unlike typical coursework, my Python pipeline allowed me to **handle 16,000 simulations algorithmically**, enabling insights beyond PK-Sim's built-in functionality.
3. **Beyond-Syllabus Computational Strategy:** Most applications focus on a **single-tool approach**. My hybrid strategy—**PBPK modeling, data engineering, validation, and visualization**—reflects industry-grade pharmacokinetic workflows.

By designing a **multi-tiered computational pipeline** that enhances PBPK modeling and data analysis, this application qualifies as a **5-level demonstration** of #CS-Computationaltools.

LO: #CS110-PythonProgramming

Description of the LO from the course syllabus: Writes Python programs to implement, analyze, and compare algorithms and apply data structures. When appropriate, produce Python code to plot and visualize meaningful performance metrics.

What is a “4” level application of this LO as described in the course? A 4-level application demonstrates correct implementation of required functionality, strategic test case design, and thorough validation. It also includes simulations and Python-based visualizations to analyze and compare algorithmic performance. A strong application would show correctness through explicit test cases and leverage Python’s scientific computing libraries to generate insightful visualizations.

What would be a “4” level application of this LO in THIS project? A 4-level application in this project would involve writing Python programs that **process, analyze, and visualize large-scale pharmacokinetic data**, ensuring the correct implementation of statistical transformations, dose-response relationships, and inter-population comparisons. Additionally, it would include well-structured test cases verifying data integrity, accuracy, and reproducibility.

Where is this LO applied in your project? This LO is applied in:

- **Appendix D:** Contains Python scripts used for pharmacokinetic modeling and data visualization.
- **Methods:** Discusses Python’s role in pre-processing and statistical validation.
- **Results:** Includes dose-stratified C_{\max} plots, placebo-drug response comparisons, and dietary sensitivity analyses.

How have I applied this LO in THIS project? Evaluate the strength of my application: **Application 1: Algorithmic Data Processing**

I wrote Python scripts to **aggregate, restructure, and normalize 16,000 pharmacokinetic simulations**. This required applying **vectorized NumPy operations** and **pandas transformations** to efficiently handle large datasets while preserving dose-response relationships.

Application 2: Validation with Test Cases

To ensure correctness, I implemented test cases that **cross-validated PK-Sim outputs**

with Python-generated results, ensuring no discrepancies in pharmacokinetic parameters like C_{\max} , T_{\max} , and clearance rates.

Application 3: Visualization of Pharmacokinetic Trends

I used **Matplotlib** and **Seaborn** to create **custom visualization techniques**, including **dose-normalized C_{\max} distributions**, **stratified plasma concentration curves**, and placebo-drug response overlays, providing analytical insights beyond PK-Sim's built-in tools.

The way I developed **an automated computational pipeline for pharmacokinetic analysis**, makes my application a **4-level standard** for #CS110-PythonProgramming.

LO: #NS164-ResearchMethods

Description of the LO from the course syllabus: Conduct scientific research; analyze experimental protocols and results. This LO focuses on designing and executing scientific experiments, analyzing protocols, and interpreting results with rigor, precision, and critical evaluation. It involves following established methodologies, systematically collecting data, and assessing the validity and reliability of both the approach and the outcomes.

What is a “4” level application of this LO as described in the course? A **4-level** application of this LO demands more than just correctly executing experimental protocols—it requires **rigorous justification** of each methodological choice. A strong application critically evaluates the strengths and limitations of different approaches, ensuring that the study design is not just functionally correct but also scientifically sound. For instance, in my previous coursework, I achieved a **4-level** application by assessing competing methodologies in molecular biology experiments, discussing how their differences affected data reliability. Similarly, in this project, achieving a high-level application requires going beyond just running pharmacokinetic simulations—I must demonstrate why each model choice, parameterization, and validation step was the most appropriate.

Where is this LO applied in my project? My application of this LO is deeply embedded in my **methodology, data validation, and comparative analysis**. From the outset, I carefully justified why **PK-Sim** was the best platform for YCT-529 pharmacokinetics, comparing it against alternative modeling tools like BioGears. Beyond just citing its features, I explained how its **whole-body physiologically based pharmacokinetic (PBPK) framework** was better suited for capturing systemic drug interactions

than compartmental models.

But research methodology is more than just picking the right tools—it’s also about ensuring that those tools produce **reliable, reproducible, and interpretable data**. That’s why I didn’t stop at running PK-Sim simulations. To validate my model, I **cross-checked its plasma concentration-time profiles using Python-based statistical analysis**, confirming consistency across different simulation outputs. I even compared my results with **Atazanavir**, a drug with a well-documented pharmacokinetic profile, to ensure my modeling choices produced realistic estimates (see).

Another major application of this LO was in how I structured my dataset for analysis. PK-Sim’s raw output wasn’t in a form that allowed for **easy cross-cohort comparison**, so I developed a data pre-processing pipeline to restructure simulation results into an aggregated format. This allowed me to examine dose-dependent trends, inter-population differences, and key pharmacokinetic parameters like **C_{\max} and clearance rates** in a way that the original software couldn’t directly visualize.

What takes this beyond a 4 and into a 5? A **typical application** of this LO would stop at well-justified modeling and validation. However, my work extends beyond standard expectations in three key ways:

1. **Mechanistic Exploration:** Rather than treating drug elimination as a simple first-order process, I investigated whether **enterohepatic recirculation** contributed to YCT-529’s **40 µg/L plateau beyond hour 10**. Identifying this as a potential explanation required integrating pharmacokinetic modeling with mechanistic hypotheses—an approach not typically expected in coursework.

2. **Application to Dose Optimization:** Instead of merely analyzing pharmacokinetics, I connected my findings to **practical dose adjustments**. By proposing **race-specific dosing strategies**, I translated raw computational results into actionable insights that could inform clinical decision-making. This level of **applied pharmacokinetic analysis** is more aligned with industry research than standard undergraduate coursework.

3. **Multi-Platform Validation:** Most pharmacokinetic modeling studies rely on a single platform. I enhanced methodological robustness by validating PK-Sim’s results using a Python-based secondary analysis and cross-referencing trends against reference drug models. This step ensured that my conclusions were not just internally consistent but also **externally validated**.

Final Reflection This project wasn’t just about running simulations—it was about **designing, validating, and critically analyzing** a pharmacokinetic model to derive

meaningful conclusions. By integrating **mechanistic hypotheses, practical dose optimization, and cross-platform validation**, my research surpasses standard coursework expectations. Given the **depth, complexity, and novel application** of my approach, this work qualifies as a **5-level demonstration** of #NS164-ResearchMethods.

LO: #ns164-modelselection

Description of the LO from the course syllabus: Evaluate the appropriate clinical and pre-clinical model for testing therapeutic applications. This LO requires selecting a model type—**in vitro, in vivo, in silico, or ex vivo**—and providing a well-reasoned justification. A strong application involves not only identifying an appropriate model but also discussing the trade-offs of different approaches. A 5-level application extends beyond conventional comparisons, offering a **novel perspective** or methodology in model selection.

What is a “4” level application of this LO as described in the course? A 4-level application requires correctly identifying the relevant model type, explaining its suitability, and comparing alternative models with a detailed discussion of their strengths and weaknesses. For instance, selecting an **in silico PBPK model** over **in vitro hepatocyte assays** based on systemic applicability would meet this standard. A thorough justification of why one approach was superior—supported by literature, known pharmacokinetic properties, and experimental feasibility—would strengthen the application.

What makes this a “5” level application in THIS project? A 5-level application needed to go beyond selecting PK-Sim as the best **in silico** tool. I framed my approach as a **hybrid validation pipeline**, integrating computational pharmacokinetics with independent Python-based data processing. Instead of relying solely on PK-Sim’s built-in tools, I **cross-validated simulated outputs**, ensuring internal consistency across multiple computational platforms. Additionally, I leveraged a **simulated placebo arm (MCC)** within a PBPK framework, an unconventional but effective way to assess YCT-529-specific pharmacokinetics.

Where is this LO applied in this project? This LO is central to my **Methodology**, particularly the sections on model selection and validation strategy. I justified my choice of PK-Sim over other pharmacokinetic simulators like **GastroPlus** or **BioGears**, showing how each tool differs in enzyme modeling, transporter integration, and physiological scaling. It also appears in the **Limitations**, where I acknowledge the trade-offs of PBPK modeling and propose experimental validation pathways.

How have I applied this LO in THIS project? Evaluate the strength of my application: **Application 1: Strategic Model Selection**

Selecting a PBPK model over conventional **in vitro** or **in vivo** studies required careful consideration. **In vivo pharmacokinetic profiling** would have provided empirical data but was impractical due to ethical and logistical constraints. While **in vitro enzyme kinetics** assays could estimate metabolism rates, they lacked systemic relevance. I chose an **in silico PBPK framework**, which allowed me to simulate YCT-529's absorption, metabolism, and elimination across diverse populations while incorporating key physiological parameters.

Application 2: Multi-Tiered Computational Validation

Rather than solely relying on PK-Sim, I conducted a **Python-based validation** to verify model reliability. By reanalyzing dose-stratified **C_{max} trends**, placebo-drug response comparisons, and inter-population variability, I ensured my conclusions were not dependent on a single computational environment. This multi-tiered verification reinforced the robustness of my results.

Application 3: Novel Use of a Simulated Placebo Control

Most PBPK studies do not incorporate **placebo pharmacokinetics**, yet I simulated **MCC** as an inert control to establish a baseline for systemic retention. This approach eliminated confounding factors, allowing me to isolate true drug absorption effects. By integrating a **placebo arm within an in silico model**, I extended my methodological design beyond conventional pharmacokinetic workflows.

Consideration for a 5: Novel Application Beyond the Course Scope This project surpasses standard coursework expectations in three critical ways:

- **Cross-Validation Across Platforms:** I did not just rely on PBPK simulations; I implemented a **secondary analytical framework** to validate predictions.
- **Interdisciplinary Model Design:** Instead of focusing solely on pharmacokinetics, I integrated **data engineering and validation techniques**, mimicking real-world pharmacometric research.
- **Innovative Model Implementation:** By embedding a **placebo arm** within a computational framework, I demonstrated a methodological advance in **in silico pharmacokinetics**.

By designing a **hybrid PBPK validation strategy** that enhances computational pharmacokinetics, this application qualifies as a **5-level demonstration** of #ns164-modelselection.

HC Appendix

HC: #thesis

Description of the HC from the course syllabus: Formulate a well-defined thesis. A strong thesis serves as the central guiding argument for any research project, ensuring clarity, coherence, and logical progression. A well-crafted thesis is **concise, arguable, substantial, precise, sets up forthcoming evidence, and relevant** (CASPER). Mastery of this HC requires the ability to construct a thesis that effectively frames the research, integrates supporting claims, and evolves based on findings.

Previous Challenges and Improvements: In an earlier draft, I received feedback indicating that my thesis statement was overly prominent within the paper’s structure, disrupting the natural flow of my argument. The critique pointed out that while my thesis was strong, its placement in the introduction did not align with academic conventions. After discussing this with my professor, I made a strategic decision to **relocate the thesis to the Appendix**. This adjustment allowed me to retain the clarity of my central argument without compromising the professional structure of my paper.

How I Applied This HC in My Project: Application 1: Developing a Precise, Arguable, and Evidence-Based Thesis

I ensured that my thesis clearly articulated the central claim of my research: **Computational pharmacokinetics provides an effective and scalable approach to optimizing YCT-529’s dosing across diverse populations**. This claim is **precise**, as it defines the scope of my work; **arguable**, as it presents a stance that can be tested against alternative approaches; and **evidence-based**, as I validate it through simulations and comparative pharmacokinetic analysis.

Application 2: Structuring the Thesis to Guide the Research

Throughout the project, my thesis evolved to align with emerging findings. Initially, I hypothesized that population-level variability would necessitate minor dosing adjustments, but as the research progressed, I found **significant inter-population differences in drug metabolism**. This led to a refined thesis that emphasizes **tailored pharmacokinetic modeling** for personalized dosing. This iterative refinement demonstrates a deep engagement with the research process and ensures that the thesis remains aligned with the actual results.

Application 3: Integrating the Thesis Without Disrupting Paper Structure

Rather than forcing the thesis into the introduction, I strategically moved it to the **Ap-**

pendix, allowing the introduction to flow naturally into the research background. This decision maintained academic integrity while still ensuring that my thesis was explicitly defined. This approach demonstrates an understanding of formal research structures and a flexible application of academic writing conventions.

Evaluation of My Application: By developing a **concise yet substantial thesis**, refining it based on findings, and strategically placing it in a manner that enhances readability, I have applied this HC at a **4-level proficiency**. However, the **adaptability in structure**, combined with the **methodological rigor in refining my claim**, suggests a creative and effective approach that aligns with a **5-level application**.

HC: #interventionalstudy

Description of the HC from the course syllabus: Design and interpret experimental studies. A true experiment involves independent variables that are manipulated, dependent variables that are measured, and the control of extraneous factors to establish causal relationships. Unlike observational studies, interventional studies actively introduce treatments or modifications to examine their effects.

What is a “4” level application of this HC as described in the course? A **4-level application** involves designing a rigorous, well-controlled experimental study where variables are explicitly justified, confounding factors are accounted for, and the results allow for strong causal inferences. This includes clear experimental design, appropriate control groups, and a discussion of extraneous influences. Additionally, study limitations and potential refinements should be identified.

What would be a “5” level application of this HC in THIS project? A **5-level application** extends beyond traditional experimental design by using novel methodologies or framing interventional studies in a unique way. In this project, I **simulate an interventional study** through PBPK modeling, where YCT-529 is “administered” computationally across **16,000 simulated individuals** at different doses and under varying conditions. This differs from traditional wet-lab experiments but still **tests hypotheses, controls variables, and evaluates outcomes**, making it a high-level intervention.

Additionally, I compare YCT-529 pharmacokinetics to **validated clinical data for Atazanavir**, treating Atazanavir as a **benchmark control** for evaluating model accuracy. This cross-validation approach ensures **real-world relevance** of computational interventions, a rarely explored extension of interventional study design.

Where is this HC applied in your project? This HC is primarily applied in the **Methodology** section, where I detail my **computational intervention approach** via PBPK modeling. It also appears in the **Discussion**, where I critically analyze the intervention outcomes against real-world pharmacokinetic data.

How have I applied this HC in this project? Evaluate the depth of application: **Application 1: Computational Drug Intervention via PBPK Modeling** I designed an **in silico** interventional study where YCT-529 was tested at different doses and across demographic groups using PK-Sim. Unlike a static dataset analysis, this approach **actively manipulates variables (dose, population, diet conditions)** to observe pharmacokinetic responses, akin to a clinical intervention but without human subjects.

Application 2: Benchmarking Against Clinical Data for Atazanavir To assess intervention validity, I compared YCT-529 pharmacokinetics to Atazanavir, a drug with established CYP3A4 metabolism. This serves as a **positive control**, ensuring my simulated interventions align with known pharmacokinetic behaviors.

Application 3: Race-Specific Dose Optimization as an Experimental Outcome Beyond simply testing a drug intervention, I used results to propose **race-specific dose modifications**, similar to precision medicine approaches for drugs like tacrolimus and warfarin. This extends interventional study design beyond a single-variable analysis to a **multi-factorial dose-response investigation**.

Consideration for a 5: This project **redefines the scope of an interventional study** by demonstrating that **computational interventions** can serve as preliminary trials before human testing. The integration of **PBPK modeling, benchmark comparisons, and race-specific interventions** represents a **novel, creative extension of traditional interventional study design**, justifying a 5-level application.

HC: #studyreplication

Description of the HC from the course syllabus: Evaluate and incorporate replicability in empirical study designs. Study replication ensures the validity of scientific findings by reproducing experimental results under similar or varied conditions. It strengthens confidence in conclusions by assessing whether observed effects persist across different methods, populations, or analytical approaches.

What is a “4” level application of this HC as described in the course? A **4-level application** demonstrates a deep understanding of **study replication by de-**

signing or executing a replication study to validate results. This includes **identifying sources of variability**, ensuring consistency in experimental methods, and critically assessing whether the replicated findings confirm or challenge the original study. The application is **well-justified, technically precise, and free from methodological gaps**. A strong example would be reanalyzing an existing dataset using independent statistical methods or **recreating a pharmacokinetic model to verify its predictive accuracy**.

What would be a “4” level application of this HC in THIS project? In this project, a **4-level application** involves **replicating Atazanavir pharmacokinetics within my PK-Sim model to validate the accuracy of my computational framework**. By lifting Atazanavir’s compound properties and designing an independent simulation protocol, I was able to **test whether my modeled enzymatic and clearance pathways produced results consistent with clinical data**. A successful replication indicated that **my metabolic assumptions, population scaling, and enzymatic kinetics were properly parameterized**, thereby increasing confidence in the model’s ability to predict YCT-529 pharmacokinetics.

Where is this HC applied in your project? This HC is primarily applied in the **Methods and Validation sections**, where I describe my approach to simulating Atazanavir as a benchmark compound. The replication process played a **critical role in confirming that the systemic clearance, C_{\max} , and elimination kinetics in my PBPK model were physiologically realistic**. Additionally, this HC is reflected in the **Results section**, where I compare the Atazanavir simulation outcomes against clinical data to demonstrate model validity.

How have I applied this HC in this project? One of the core challenges in PBPK modeling is ensuring that simulated drug behavior aligns with **known, real-world pharmacokinetics**. I applied **study replication** by independently running an Atazanavir simulation, comparing my results to published clinical data. The **close alignment between my simulated and observed concentration-time profiles** confirmed that my model accurately captured **CYP3A4 metabolism, biliary excretion, and systemic drug clearance**.

Additionally, **this replication served as an internal consistency check**. If Atazanavir—a drug with well-characterized PK parameters—had yielded unrealistic simulation results, it would have flagged potential **errors in my model’s enzyme kinetics or clearance rates**. By verifying that Atazanavir behaved as expected, I demonstrated

a rigorous approach to validating my model before applying it to YCT-529 predictions.

Lastly, my study replication allowed me to **fine-tune my PK-Sim parameterization strategies**. For example, observing Atazanavir’s clearance patterns informed my **population variability assumptions**, which were then applied to refining YCT-529 simulations. This illustrates how **replicating existing pharmacokinetic models enhances confidence in novel predictions**, a critical component of **computational pharmacology and translational drug development**.

HC: #evidencebased

Description of the HC from the course syllabus: Identify and appropriately structure the information needed to support an argument effectively. This HC requires selecting compelling, well-cited evidence that aligns with the argument, structuring it for clarity, and addressing counterarguments or contradictions to reinforce credibility.

What is a “4” level application of this HC as described in the course? A **4-level application** entails not only presenting well-supported arguments but also **structuring evidence logically, addressing potential counterarguments, and ensuring every claim is backed by substantial, high-quality evidence**. This includes carefully choosing between qualitative and quantitative evidence, using **data visualizations where appropriate**, and ensuring that the presented evidence strengthens the argument.

What would be a “4” level application of this HC in THIS project? In this project, a **4-level application** involves integrating pharmacokinetic data, enzymatic metabolism pathways, and comparative drug studies to support claims regarding YCT-529’s absorption, distribution, metabolism, and elimination. This means **using clinical data from Atazanavir simulations as a benchmark**, referencing **PBPK modeling literature**, and aligning computational findings with known metabolic principles. Additionally, figures such as **plasma concentration-time profiles** serve as direct evidence to illustrate trends in systemic exposure.

Where is this HC applied in your project? This HC is most evident in the **Results and Discussion sections**, where I analyze plasma concentration trends and enzymatic interactions. It is also applied in the **Methodology section**, where I justify parameter selection using literature-backed evidence. Throughout the paper, **Vancouver-style citations** ensure proper attribution and alignment with scientific publishing norms.

How have I applied this HC in this project? One key application of this HC is in **validating my PBPK model**. To ensure reliability, I **simulated Atazanavir pharmacokinetics independently**, then compared the results to published clinical data. The strong alignment between simulated and real-world data provided compelling evidence that **my enzymatic and clearance parameters were well-calibrated**.

Another critical application is **structuring comparative evidence to justify race-specific dosing strategies**. By analyzing how YCT-529's C_{\max} varied across populations, I argued that **East Asians require lower doses due to reduced CYP3A4 activity, while Black Americans may need higher doses**. This claim was substantiated using literature on **racial pharmacokinetic variability in tacrolimus and warfarin**.

Lastly, I applied **evidence-based reasoning in dietary condition analysis**, demonstrating how **food intake affects systemic exposure**. Fasting led to higher C_{\max} , a pattern supported by studies on **bile-mediated drug solubilization**. These examples illustrate **a well-structured, data-driven approach**, making this a strong application of #evidencebased.

HC: #professionalism

Description of the HC from the course syllabus: Follow established guidelines to present yourself and your work products professionally. This HC emphasizes adherence to field-specific norms, ensuring clarity, accuracy, and credibility in communication.

What would be a “4” level application of this HC in THIS project? A 4-level application requires strict adherence to professional formatting standards, including proper citation styles, structured documentation, and clear, reproducible presentation of research.

How is this HC applied in my project? I demonstrated **#professionalism** by using **Vancouver citation style**, the standard in pharmacokinetics, ensuring my work aligns with scientific conventions. My GitHub repository includes **structured documentation, dataset access instructions, and clear computational methodology**, reinforcing transparency and reproducibility.

I formatted **pharmacokinetic tables** with consistent units, clear headings, and readable columns. Additionally, I refined **figures and graphs** to ensure clarity and accuracy. Consulting my professor, I moved my thesis statement to the appendix, preserving the research structure while maintaining clarity.

By rigorously maintaining professional standards in writing, citation, and data presentation, I demonstrate a strong **4-level application** of #professionalism.